

Lecture 5

Image Characterization

ch. 4 of *Machine Vision* by Wesley E. Snyder & Hairong Qi

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16-725 (CMU RI) : BioE 2630 (Pitt)

Dr. John Galeotti



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Digital Images

- How are they formed?
- How can they be represented?

Image Representation

- Hardware
 - Storage
 - Manipulation
- Human
 - Conceptual
 - Mathematical

Iconic Representation

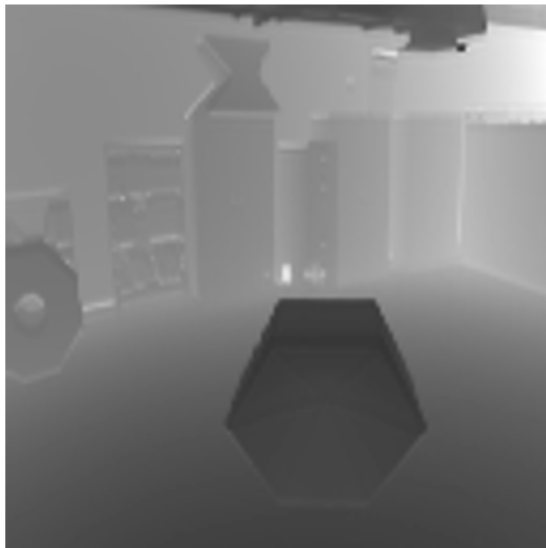
- What you think of as an image, ...
 - Camera
 - X-Ray
 - CT
 - MRI
 - Ultrasound
 - 2D, 3D, ...
 - etc



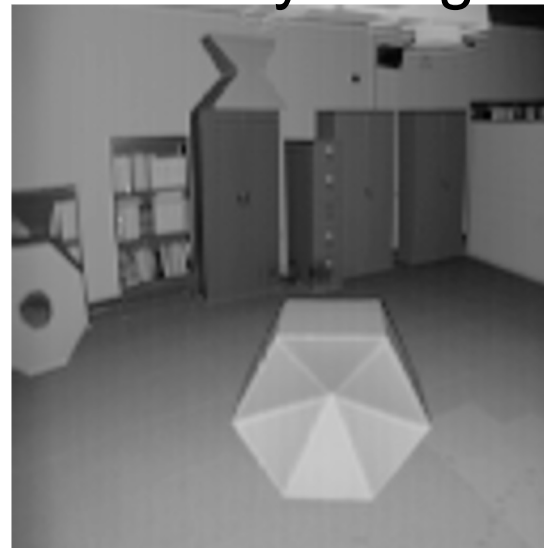
Iconic Representation

- And what you might not

Range Image



Corresponding Intensity Image



Images from CESAR lab at Oak Ridge National Laboratory,
Sourced from the USF Range Image Database:
<http://marathon.csee.usf.edu/range/DataBase.html>
Acknowledgement thereof requested with redistribution.

Functional Representation

- An Equation

- Typically continuous

- Fit to the image data

- Sometimes the entire image
- Usually just a small piece of it

- Examples (Quadratic Surfaces):

- Explicit: $z = ax^2 + by^2 + cxy + dx + ey + f$

- Implicit: $0 = ax^2 + by^2 + cz^2 + dxy + exz + fyz + gx + hy + iz + j$

Linear Representation

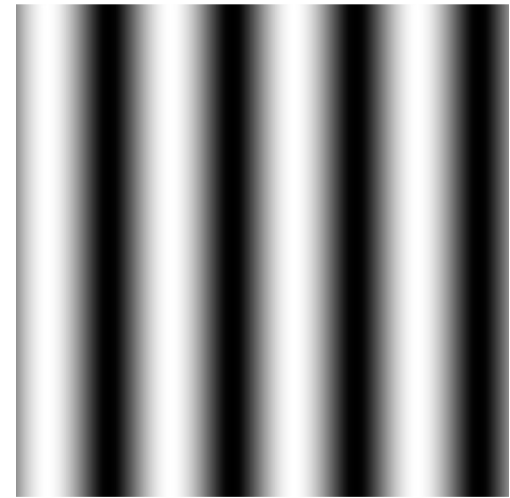
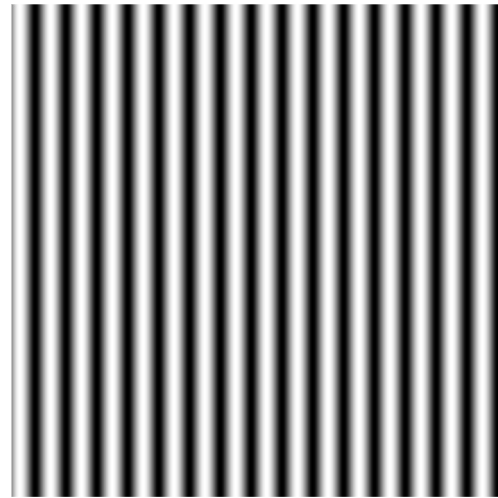
- Unwind the image
 - “Raster-scan” it
- Entire image is now a vector
 - Now we can do matrix operations on it!
 - Often used in research papers

Probabilistic & Relational Representations

- Probability & Graphs
- Discussed later (if at all)

Spatial Frequency Representation

- Think “Fourier Transform”
- Multiple Dimensions!
- Varies greatly across different image regions
- High Freq. = Sharpness



- Steven Lehar’s details:
http://sharp.bu.edu/~sl_ehar/fourier/fourier.html

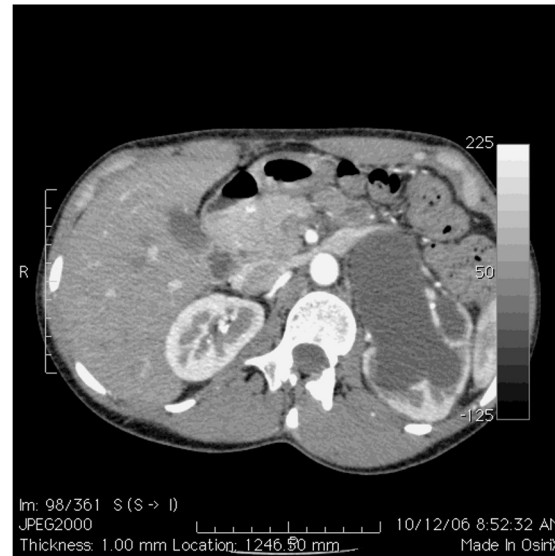
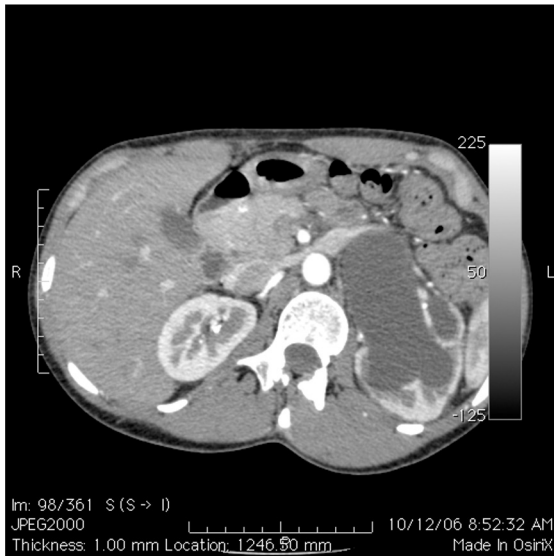


Image Formation

- Sampling an analog signal
- Resolution
 - # Samples per dimension, OR
 - Smallest clearly discernable physical object
- Dynamic Range
 - # bits / pixel (quantization accuracy), OR
 - Range of measurable intensities
 - Physical meaning of min & max pixel values
 - light, density, etc.

Dynamic Range Example

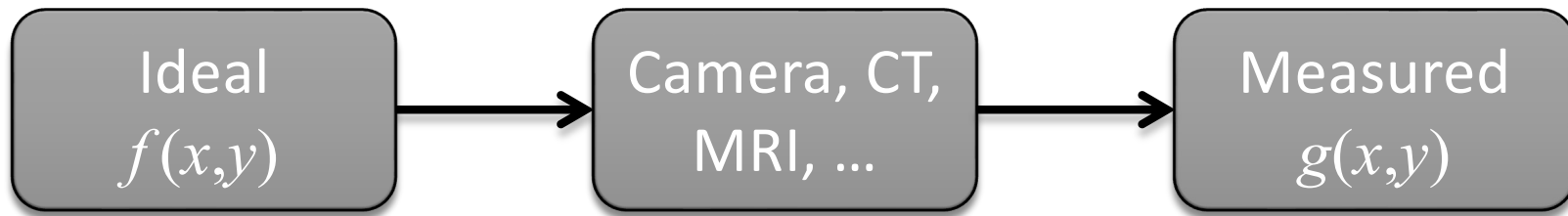
(A slice from a Renal Angio CT: 8 bits, 4 bits, 3 bits, 2 bits)



An Aside: The Correspondence Problem

- My Definition:
 - Given two different images of the same (or similar) objects,
for any point in one image
determine the exact corresponding point in the other image
- Similar (identical?) to registration
- Quite possibly, it is THE problem in computer vision

Image Formation: Corruption



- There is an ideal image
 - It is what we are physically measuring
- No measuring device is perfect
 - Measuring introduces noise
 - $g(x,y) = D(f(x,y))$, where D is the distortion function
- Often, noise is additive and independent of the ideal image

Image Formation: Corruption

- Noise is usually not the only distortion
- If the other distortions are:
 - linear &
 - space-invariant

then they can *always* be represented with the convolution integral!

- Total corruption:

$$g(x, y) = \iint_{-\infty \dots \infty} f(\alpha, \beta) h(x - \alpha, y - \beta) d\alpha d\beta + n(x, y)$$

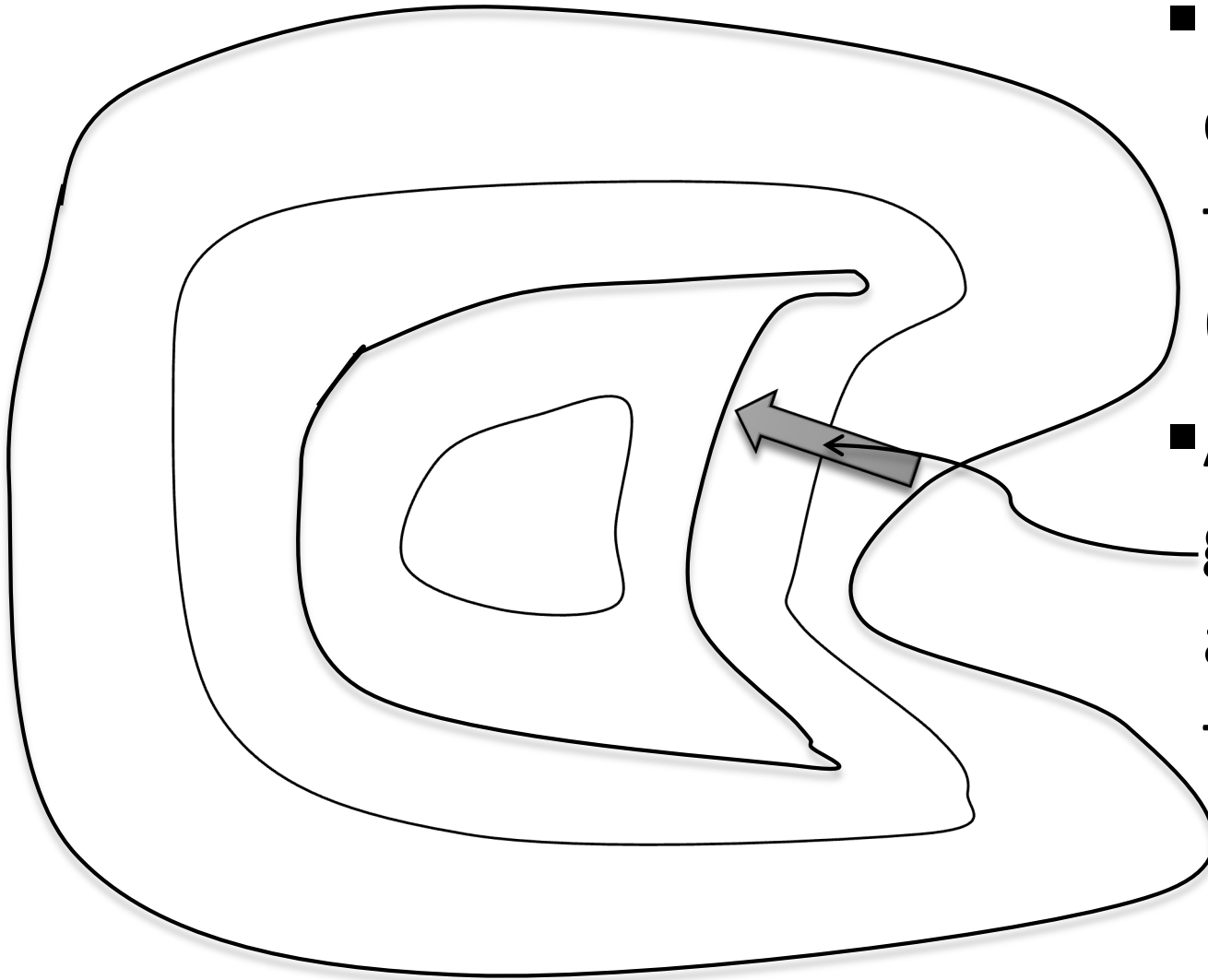
The image as a surface

- Intensity \rightarrow height
 - In 2D case, but concepts extend to ND
- $z = f(x, y)$
- Describes a surface in space
 - Because only one z value for each x, y pair
 - Assume surface is continuous (interpolate pixels)

Isophote

- “Uniform brightness”
- $C = f(x, y)$
- A curve (2D) or surface (3D) in space
- Always perpendicular to image gradient
 - Why?

Isophotes & Gradient



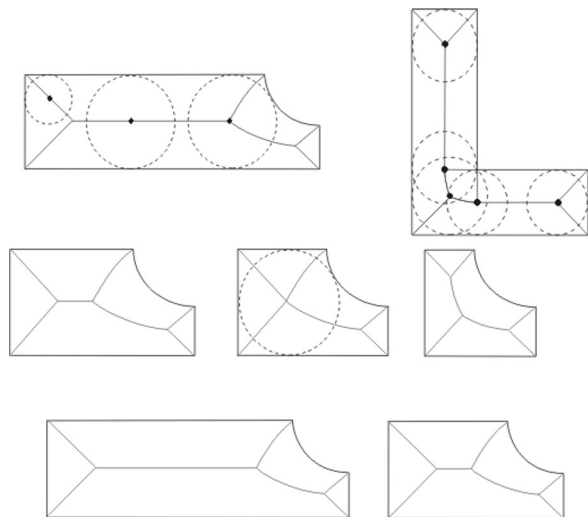
- Isophotes are like contour lines on a topography (elevation) map.
- At any point, the gradient is always at a right angle to the isophote!

Ridges

- One definition:
 - Local maxima of the rate of change of gradient direction
 - Sound confusing?
 - Just think of ridge lines along a mountain
 - If you need it, look it up
 - Snyder references Maintz

Medial Axis

- Skeletal representation
- Defined for binary images
 - This includes segmented images!
- “Ridges in scale-space”
 - Details have to wait (ch. 9)



<http://sog1.me.qub.ac.uk/Research/medial/medial.php>

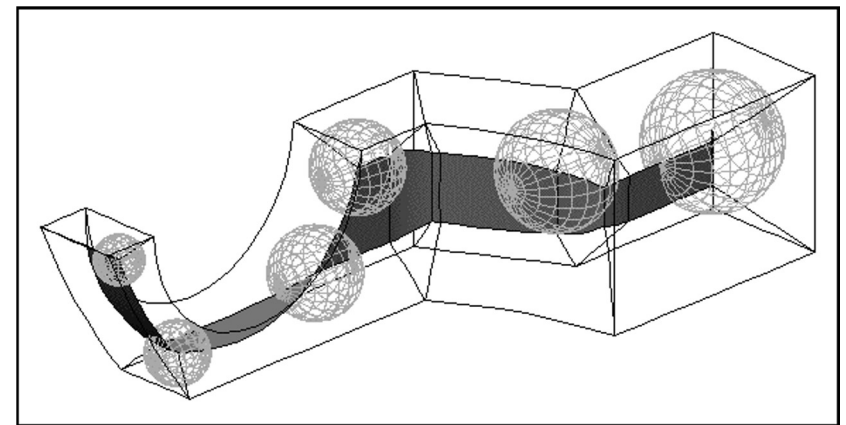
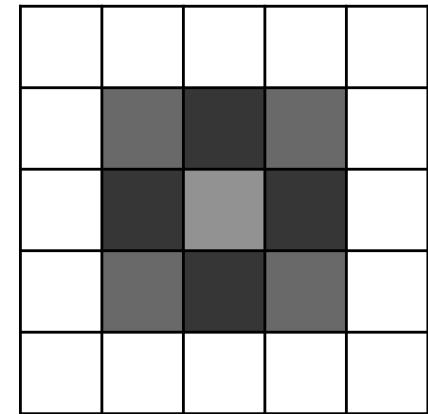


Image courtesy of TranscenData Europe
<http://www.fegs.co.uk/motech.html>

Neighborhoods

- Terminology

- 4-connected vs. 8-connected
- Side/Face-connected vs. vertex-connected
- Maximally-connected vs. minimally-connected (ND)



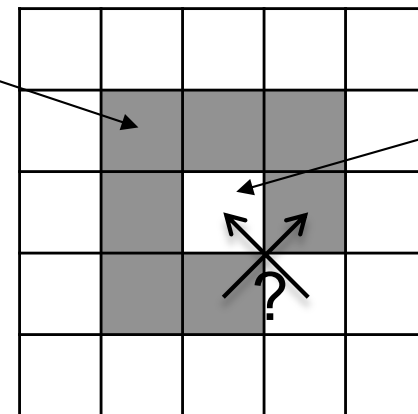
- Connectivity paradox

- Due to discretization

- Can define other neighborhoods

- Adjacency not necessarily required

Is this shape closed?



Is this pixel connected to the outside?

Curvature

- Compute curvature at every point in a (range) image
 - (Or on a segmented 3D surface)
- Based on differential geometry
- Formulas are in your book
- 2 scalar measures of curvature that are invariant to viewpoint, derived from the 2 principal curvatures, (K_1, K_2) :
 - Mean curvature (arithmetic mean)
 - Gauss curvature (product)
 - =0 if either $K_1=0$ or $K_2=0$